TheAnalyticsTeam

# Sprocket Central Pty Ltd

Data analytics approach

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## Agenda

- 1. Introduction
- 2. Data Exploration
- 3. Model Development
- 4. Interpretation

## Introduction

Based on the dataset provided, detail approach for following slides. Our team focus:

- 1. Understanding Data Distribution
- 2. Feature Engineering
- 3. Data Transformation
- 4. Model building
- 5. Interpretation

### **Understanding Data:**

#### **Dataset Customer Demographic:**

- 1 Total rows are 4000 and 13 columns
- 2. Missing values: Last Name 125 rows with missing last name, 87 rows with missing DOB, 506 missing job title, 656 missing industry category, 87 rows with missing tenures.
- 3. Inconsistency in values in gender columns, there are values with M,F, Male, Female and U

#### **Dataset Customer Address:**

- The dataset has no missing values.
- 2. The state defined is not consistent, state is defined as NSW and also New South Wales.

#### **Dataset Transaction:**

- 1. There are 20,000 records of transaction for year 2017.
- 2. Missing values: Column online order is missing 360 records with no values.
- 3. Missing values: Columns brand, product\_line, product\_class, product\_size, standard\_cost are missing values for 197 records. If records is missing one column the same row has missing value in other above columns.

#### **Feature Engineering Dataset:**

Merge Datasets: Customer Demographic, Customer Address and Transactions:

- 1. Dropping row element with extreme DOB value: 1843-12-21, as it is skewed data.
- 2. Rename Gender values and some values were F/M. Renaming U as 'Unknown'
- 3. Replacing missing values with 'Unknown' in Job Title and Job Industry Category.
- 4. Deceased Indicator has most values as N, dropping the column.
- Created additional column Age using DOB.
- 6. State columns has ambiguous values, Victoria and New South Wales, converting them to VIC and NSW respectively.

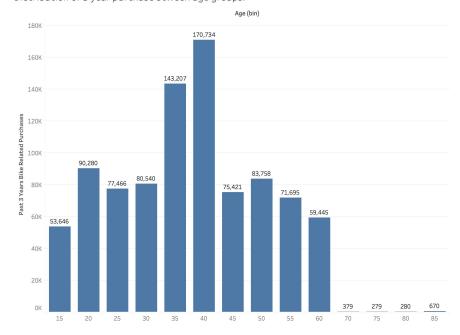
#### Analysis on Transaction Dataset:

- 7. Similar values are missing in columns: Brand, Product Line, Product Class and Product Size, dropping these rows.
- 8. Created additional column Profit using List Price and Standard Cost
- 9. Created additional column Age using DOB.

#### **Customer Segmentation based on Age:**

Distribution of people based on age groups, we can observe from the data that total number of rental purchases of bike for past 3 years are between age group 30 and 45.

Distribution of 3 year purchase btween age groups.

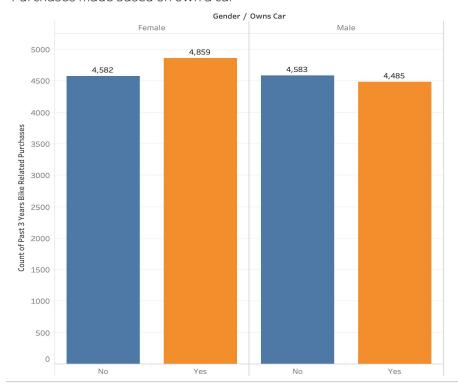


#### **Customer Segmentation based on Gender:**

Distribution of people based on gender group, we can observe from the data of past 3 years of purchases:

- 1. Females are more likely to rent the bike than men, females have rented total 9,441 times compared to men 9,068 times.
- 2. Also, females with cars are more likely to rent a bike than men, females with cars have rented total of 4,859 to men 4,485.
- 3. There are no difference in rental of bikes if they don't own a car.

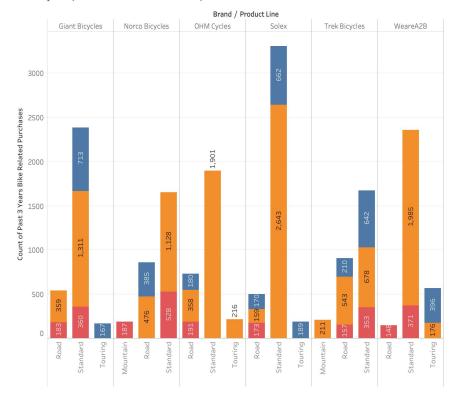
#### Purchases made based on own a car



## **Brand Segmentation based on 3 year rental:**

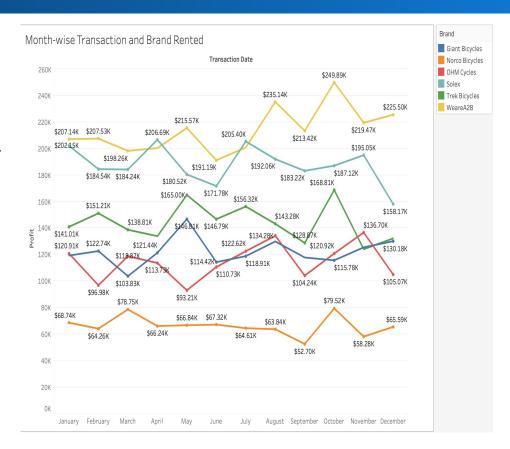
- 1. The most popular brand among the people is Solex Standard Product line with total of 3,305 times rented.
- 2. Overall for all the brands standard is most popular product line.

Past 3 year purchases for brand and product line.



#### **Monthwise distribution Profits:**

- 1. The most profitable brand is WeareA2B.
- 2. The most profitable month is October for for year 2017 with sales profit of \$922.03K.



## **Hotspots Area of most Profitable area:**

1. New South Wales and Victoria Area is more profitable rental area for bikes.

